

Creating Linear Regression Model Using PySpark

Install PySpark

My Roll No. Is : DS5B-2137

```
In [ ]: pip install pyspark
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Collecting pyspark
  Downloading pyspark-3.2.1.tar.gz (281.4 MB)
    |████████████████████████████████████████| 281.4 MB 34 kB/s
Collecting py4j==0.10.9.3
  Downloading py4j-0.10.9.3-py2.py3-none-any.whl (198 kB)
    |████████████████████████████████████████| 198 kB 46.5 MB/s
Building wheels for collected packages: pyspark
  Building wheel for pyspark (setup.py) ... done
  Created wheel for pyspark: filename=pyspark-3.2.1-py2.py3-none-any.whl size=281853642 sha256=99657e37a6edb52a83d4b4e280e11c4e26947120a4b2dc8192749712f371238e
  Stored in directory: /root/.cache/pip/wheels/9f/f5/07/7cd8017084dce4e93e84e92efd1e1d5334db05f2e83bcef74f
Successfully built pyspark
Installing collected packages: py4j, pyspark
Successfully installed py4j-0.10.9.3 pyspark-3.2.1
```

Import Library and Creating Session

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```
In [ ]: from pyspark.sql import SparkSession
```

```
In [ ]: session = SparkSession.builder.appName("exam1").master("local").getOrCreate()
```

Read Dataset

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```
In [ ]: data = session.read.csv("Big Mart Sale.csv", header = True, inferSchema=True)
```

To print top 10 raw in dataset

```
In [ ]: data.show(10)
```

```
+-----+-----+-----+-----+-----+-----+-----+
--+-+-----+-----+-----+-----+-----+-----+-----+
-----+-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_
MRP|Outlet_Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|O
utlet_Type|Item_Outlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+
--+-+-----+-----+-----+-----+-----+-----+-----+
```

```

+-----+
|          FDA15|          9.3|          Low Fat|          0.016047301|          Dairy|249.8
092|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          3735.138|
|          DRC01|          5.92|          Regular|          0.019278216|          Soft Drinks| 48.2
692|          OUT018|          2009|          Medium|          Tier 3|Superma
rket Type2|          443.4228|
|          FDN15|          17.5|          Low Fat|          0.016760075|          Meat| 141.
618|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          2097.27|
|          FDX07|          19.2|          Regular|          0.0|Fruits and Vegeta...| 182.
095|          OUT010|          1998|          null|          Tier 3|Gro
cery Store|          732.38|
|          NCD19|          8.93|          Low Fat|          0.0|          Household| 53.8
614|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          994.7052|
|          FDP36|          10.395|          Regular|          0.0|          Baking Goods| 51.4
008|          OUT018|          2009|          Medium|          Tier 3|Superma
rket Type2|          556.6088|
|          FDO10|          13.65|          Regular|          0.012741089|          Snack Foods| 57.6
588|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          343.5528|
|          FDP10|          null|          Low Fat|          0.127469857|          Snack Foods|107.7
622|          OUT027|          1985|          Medium|          Tier 3|Superma
rket Type3|          4022.7636|
|          FDH17|          16.2|          Regular|          0.016687114|          Frozen Foods| 96.9
726|          OUT045|          2002|          null|          Tier 2|Superma
rket Type1|          1076.5986|
|          FDU28|          19.2|          Regular|          0.09444959|          Frozen Foods|187.8
214|          OUT017|          2007|          null|          Tier 2|Superma
rket Type1|          4710.535|
+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+
-----+-----+
only showing top 10 rows

```

Check Null Values in columns

```
In [ ]: from pyspark.sql.functions import isnan, when, count, col
```

```
In [ ]: data.select([count(when(isnan(c) | col(c).isNull(), c)).alias(c) for c in data.columns])

+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+
-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_MRP|Outlet_
Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|Outlet_Type|Item_O
utlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+
-----+
|          0|          1463|          0|          0|          0|          0|
          0|          0|          2410|          0|          0|
          0|
+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+
-----+

```

```
In [ ]: import pyspark.sql.functions as func
```

```
In [ ]: data.agg(func.percentile_approx("Item_Weight", 0.5).alias("mean")).show()
```

```
+-----+
|mean|
+-----+
|12.6|
+-----+
```

Fill Null Values

First we replace 12.6 in place of Null values in Item_weight column because it is mean in this column

```
In [ ]: data = data.na.fill(value=12.6,subset=["Item_Weight"])
```

Second we return Medium in place of Null values in Outlet_Size Column Because Medium is the median in Outlet_Size Column

```
In [ ]: data = data.na.fill(value="Medium",subset=["Outlet_Size"])
```

```
In [ ]: data.show()
```

```
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_
MRP|Outlet_Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|
Outlet_Type|Item_Outlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|          FDA15|          9.3|          Low Fat|          0.016047301|          Dairy|249.8
092|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          3735.138|
|          DRC01|          5.92|          Regular|          0.019278216|          Soft Drinks| 48.2
692|          OUT018|          2009|          Medium|          Tier 3|Superma
rket Type2|          443.4228|
|          FDN15|          17.5|          Low Fat|          0.016760075|          Meat| 141.
618|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          2097.27|
|          FDX07|          19.2|          Regular|          0.0|Fruits and Vegeta...| 182.
095|          OUT010|          1998|          Medium|          Tier 3|Gro
cery Store|          732.38|
|          NCD19|          8.93|          Low Fat|          0.0|          Household| 53.8
614|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          994.7052|
|          FDP36|          10.395|          Regular|          0.0|          Baking Goods| 51.4
008|          OUT018|          2009|          Medium|          Tier 3|Superma
rket Type2|          556.6088|
|          FDO10|          13.65|          Regular|          0.012741089|          Snack Foods| 57.6
588|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          343.5528|
|          FDP10|          12.6|          Low Fat|          0.127469857|          Snack Foods|107.7
622|          OUT027|          1985|          Medium|          Tier 3|Superma
rket Type3|          4022.7636|
|          FDH17|          16.2|          Regular|          0.016687114|          Frozen Foods| 96.9
726|          OUT045|          2002|          Medium|          Tier 2|Superma
rket Type1|          1076.5986|
|          FDU28|          19.2|          Regular|          0.09444959|          Frozen Foods|187.8
214|          OUT017|          2007|          Medium|          Tier 2|Superma
rket Type1|          4710.535|
|          FDY07|          11.8|          Low Fat|          0.0|Fruits and Vegeta...| 45.5
402|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          1516.0266|
```

102	FDA03	18.5	Regular	0.045463773	Dairy 144.1
rket Type1	OUT046	2187.153	1997	Small	Tier 1 Superma
	FDX32	15.1	Regular	0.1000135	Fruits and Vegeta... 145.4
786	OUT049	1589.2646	1999	Medium	Tier 1 Superma
rket Type1	FDS46	17.6	Regular	0.047257328	Snack Foods 119.6
782	OUT046	2145.2076	1997	Small	Tier 1 Superma
rket Type1	FDF32	16.35	Low Fat	0.0680243	Fruits and Vegeta... 196.4
426	OUT013	1977.426	1987	High	Tier 3 Superma
rket Type1	FDP49	9.0	Regular	0.069088961	Breakfast 56.3
614	OUT046	1547.3192	1997	Small	Tier 1 Superma
rket Type1	NCB42	11.8	Low Fat	0.008596051	Health and Hygiene 115.3
492	OUT018	1621.8888	2009	Medium	Tier 3 Superma
rket Type2	FDP49	9.0	Regular	0.069196376	Breakfast 54.3
614	OUT049	718.3982	1999	Medium	Tier 1 Superma
rket Type1	DRI11	12.6	Low Fat	0.034237682	Hard Drinks 113.2
834	OUT027	2303.668	1985	Medium	Tier 3 Superma
rket Type3	FDU02	13.35	Low Fat	0.10249212	Dairy 230.5
352	OUT035	2748.4224	2004	Small	Tier 2 Superma
rket Type1					

only showing top 20 rows

A Simple Exploratory Of Dataset

To print all columns name

```
In [ ]: data.columns
```

```
Out[ ]: ['Item_Identifier',
        'Item_Weight',
        'Item_Fat_Content',
        'Item_Visibility',
        'Item_Type',
        'Item_MRP',
        'Outlet_Identifier',
        'Outlet_Establishment_Year',
        'Outlet_Size',
        'Outlet_Location_Type',
        'Outlet_Type',
        'Item_Outlet_Sales']
```

To count total numbers of rows in dataset

```
In [ ]: data.count()
```

```
Out[ ]: 8523
```

To print the schema of dataset, Spark schema is the structure of the DataFrame or Dataset, which is a collection of StructField that define the column name(String), column type (DataType), nullable column (Boolean) and metadata (MetaData)

```
In [ ]: data.printSchema()
```

```
root
 |-- Item_Identifier: string (nullable = true)
 |-- Item_Weight: double (nullable = false)
 |-- Item_Fat_Content: string (nullable = true)
 |-- Item_Visibility: double (nullable = true)
 |-- Item_Type: string (nullable = true)
 |-- Item_MRP: double (nullable = true)
 |-- Outlet_Identifier: string (nullable = true)
 |-- Outlet_Establishment_Year: integer (nullable = true)
 |-- Outlet_Size: string (nullable = false)
 |-- Outlet_Location_Type: string (nullable = true)
 |-- Outlet_Type: string (nullable = true)
 |-- Item_Outlet_Sales: double (nullable = true)
```

To know data type of each columns

```
In [ ]: data.dtypes
```

```
Out[ ]: [('Item_Identifier', 'string'),
 ('Item_Weight', 'double'),
 ('Item_Fat_Content', 'string'),
 ('Item_Visibility', 'double'),
 ('Item_Type', 'string'),
 ('Item_MRP', 'double'),
 ('Outlet_Identifier', 'string'),
 ('Outlet_Establishment_Year', 'int'),
 ('Outlet_Size', 'string'),
 ('Outlet_Location_Type', 'string'),
 ('Outlet_Type', 'string'),
 ('Item_Outlet_Sales', 'double')]
```

Data Preprocessing

Here we convert the data into machine readable form

VectorAssembler :- It is feature transformer that combine multiple columns into a single vector column.

StringIndexer :- It is use for mapping a string column to a index column that will be treated as a categorical column by spark.

OneHotEncoder :- It is an important technique for converting categorical attributes into a numeric vector

```
In [ ]: from pyspark.ml.feature import VectorAssembler, StringIndexer, OneHotEncoder
```

```
In [ ]: str_index = StringIndexer(inputCols = ['Item_Identifier', 'Item_Fat_Content', 'Item_Type',
```

```
In [ ]: one_hot = OneHotEncoder(inputCols = ['Item_Identifier1', 'Item_Fat_Content1', 'Item_Type1',
```

```
In [ ]: vector_ass = VectorAssembler(inputCols = ['Item_Weight', 'Item_Fat_Content2', 'Item_Visibi
```

Import Linear Regression and Create Model

```
In [ ]: from pyspark.ml.regression import LinearRegression
```

```
In [ ]: linear = LinearRegression(featuresCol="allfeatures", labelCol="Item_Outlet_Sales")
```

Create Pipeline for ML Model

```
In [ ]: from pyspark.ml import Pipeline
mypipeline = Pipeline(stages = [str_index, one_hot, vector_ass, linear])
```

Making Train Test Split

My Roll No. Is : 37

Using randomsplit data is split into 77% of training and 23% of test as given

```
In [ ]: training, test = data.randomSplit([0.77, 0.23])
```

Model Training

```
In [ ]: lin_reg_model = mypipeline.fit(training)
```

Test Model

```
In [ ]: result = lin_reg_model.transform(test)
```

```
In [ ]: result.show()
```

```
+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|  Item_Type|Item_MRP|Outle
t_Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|      Outlet_Type
e|Item_Outlet_Sales|Item_Identifier1|Item_Fat_Content1|Item_Type1|Outlet_Identifier1|Out
let_Establishment_Year1|Outlet_Size1|Outlet_Location_Type1|Outlet_Type1|      Item_Identifi
er2|Item_Fat_Content2|      Item_Type2|Outlet_Identifier2|Outlet_Establishment_Year2| Outl
et_Size2|Outlet_Location_Type2| Outlet_Type2|      allfeatures|      prediction|
+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+
|          DRA12|          11.6|          Low Fat|          0.0|Soft Drinks|141.6154|
OUT045|          2002|          Medium|          Tier 2|Supermarket Type
1|          3829.0158|          1051.0|          0.0|          8.0|          7.0|
          7.0|          0.0|          1.0|          0.0|(1553,[1051],[1.
0])|(4,[0],[1.0])|(15,[8],[1.0])|(9,[7],[1.0])|(8,[7],[1.0])|(2,
[0],[1.0])|(2,[1],[1.0])|(3,[0],[1.0])|(29,[0,1,14,21,22,...| 2277.372227927723|
|          DRA12|          11.6|          Low Fat|          0.041112694|Soft Drinks|142.0154|
OUT018|          2009|          Medium|          Tier 3|Supermarket Type
2|          850.8924|          1051.0|          0.0|          8.0|          5.0|
          5.0|          0.0|          0.0|          3.0|(1553,[1051],[1.
0])|(4,[0],[1.0])|(15,[8],[1.0])|(9,[5],[1.0])|(8,[5],[1.0])|(2,
[0],[1.0])|(2,[0],[1.0])|(3,[],[1.0])|(29,[0,1,5,14,21,...| 1937.56029983537|
|          DRA12|          11.6|          Low Fat|          0.068535039|Soft Drinks|143.0154|
OUT010|          1998|          Medium|          Tier 3|          Grocery Stor
```

e	283.6308	1051.0	0.0	8.0	8.0
	8.0	0.0	0.0	1.0	(1553,[1051],[1.0])
0))	(4,[0],[1.0])	(15,[8],[1.0])	(9,[8],[1.0])	(8,[],[])	(2,[0],[1.0])
	(2,[0],[1.0])	(3,[1],[1.0])	(29,[0,1,5,14,21,...	327.38010298800816	
	DRA24	19.35	Regular	0.039920687	Soft Drinks 163.3868
	OUT035	2004	Small	Tier 2	Supermarket Type
1	3439.5228	322.0	1.0	8.0	1.0
	2.0	1.0	1.0	0.0	(1553,[322],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[1],[1.0])	(8,[2],[1.0])	(2,[1],[1.0])
	(2,[1],[1.0])	(3,[0],[1.0])	(29,[0,2,5,14,21,...	2671.7706482761328	
	DRA24	19.35	Regular	0.040154087	Soft Drinks 164.6868
	OUT017	2007	Medium	Tier 2	Supermarket Type
1	1146.5076	322.0	1.0	8.0	2.0
	3.0	0.0	1.0	0.0	(1553,[322],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[2],[1.0])	(8,[3],[1.0])	(2,[0],[1.0])
	(2,[1],[1.0])	(3,[0],[1.0])	(29,[0,2,5,14,21,...	2691.662378943857	
	DRA59	8.27	Regular	0.0	Soft Drinks 183.2924
	OUT017	2007	Medium	Tier 2	Supermarket Type
1	2406.2012	97.0	1.0	8.0	2.0
	3.0	0.0	1.0	0.0	(1553,[97],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[2],[1.0])	(8,[3],[1.0])	(2,[0],[1.0])
	(2,[1],[1.0])	(3,[0],[1.0])	(29,[0,2,14,21,22...	3009.0974065421287	
	DRA59	12.6	Regular	0.127308434	Soft Drinks 186.6924
	OUT027	1985	Medium	Tier 3	Supermarket Type
3	7033.5112	97.0	1.0	8.0	4.0
	0.0	0.0	0.0	2.0	(1553,[97],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[4],[1.0])	(8,[0],[1.0])	(2,[0],[1.0])
	(2,[0],[1.0])	(3,[2],[1.0])	(29,[0,2,5,14,21,...	4393.2781578674	
	DRB01	7.39	Low Fat	0.082367244	Soft Drinks 187.753
	OUT049	1999	Medium	Tier 1	Supermarket Type
1	1518.024	1336.0	0.0	8.0	3.0
	4.0	0.0	2.0	0.0	(1553,[1336],[1.0])
0))	(4,[0],[1.0])	(15,[8],[1.0])	(9,[3],[1.0])	(8,[4],[1.0])	(2,[0],[1.0])
	(2,[],[])	(3,[0],[1.0])	(29,[0,1,5,14,21,...	2996.614390466222	
	DRB13	6.115	Regular	0.007043008	Soft Drinks 190.353
	OUT035	2004	Small	Tier 2	Supermarket Type
1	569.259	1052.0	1.0	8.0	1.0
	2.0	1.0	1.0	0.0	(1553,[1052],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[1],[1.0])	(8,[2],[1.0])	(2,[1],[1.0])
	(2,[1],[1.0])	(3,[0],[1.0])	(29,[0,2,5,14,21,...	3120.027487849694	
	DRB13	6.115	Regular	0.01179078	Soft Drinks 189.053
	OUT010	1998	Medium	Tier 3	Grocery Stor
e	948.765	1052.0	1.0	8.0	8.0
	8.0	0.0	0.0	1.0	(1553,[1052],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[8],[1.0])	(8,[],[])	(2,[0],[1.0])
	(2,[0],[1.0])	(3,[1],[1.0])	(29,[0,2,5,14,21,...	1146.3741600280819	
	DRB25	12.3	Low Fat	0.069446588	Soft Drinks 106.3938
	OUT035	2004	Small	Tier 2	Supermarket Type
1	857.5504	323.0	0.0	8.0	1.0
	2.0	1.0	1.0	0.0	(1553,[323],[1.0])
0))	(4,[0],[1.0])	(15,[8],[1.0])	(9,[1],[1.0])	(8,[2],[1.0])	(2,[1],[1.0])
	(2,[1],[1.0])	(3,[0],[1.0])	(29,[0,1,5,14,21,...	1711.1967642199536	
	DRB48	12.6	Regular	0.024733134	Soft Drinks 40.2822
	OUT027	1985	Medium	Tier 3	Supermarket Type
3	1296.3126	672.0	1.0	8.0	4.0
	0.0	0.0	0.0	2.0	(1553,[672],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[4],[1.0])	(8,[0],[1.0])	(2,[0],[1.0])
	(2,[0],[1.0])	(3,[2],[1.0])	(29,[0,2,5,14,21,...	2156.0649493401916	
	DRB48	16.75	Regular	0.024848788	Soft Drinks 39.9822
	OUT035	2004	Small	Tier 2	Supermarket Type
1	746.3618	672.0	1.0	8.0	1.0
	2.0	1.0	1.0	0.0	(1553,[672],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[1],[1.0])	(8,[2],[1.0])	(2,[1],[1.0])
	(2,[1],[1.0])	(3,[0],[1.0])	(29,[0,2,5,14,21,...	769.7994983213209	
	DRB48	16.75	Regular	0.041599644	Soft Drinks 40.9822
	OUT010	1998	Medium	Tier 3	Grocery Stor

```

e|      157.1288|      672.0|      1.0|      8.0|      8.0|
      8.0|      0.0|      0.0|      1.0| (1553,[672],[1.
0])|      (4,[1],[1.0])|(15,[8],[1.0])|      (9,[8],[1.0])|      (8,[],[])|(2,
[0],[1.0])|      (2,[0],[1.0])|(3,[1],[1.0])|(29,[0,2,5,14,21,...|-1171.6910866866936|
|      DRC01|      5.92|      Regular|      0.019278216|Soft Drinks| 48.2692|
      OUT018|      2009|      Medium|      Tier 3|Supermarket Type
2|      443.4228|      673.0|      1.0|      8.0|      5.0|
      5.0|      0.0|      0.0|      3.0| (1553,[673],[1.
0])|      (4,[1],[1.0])|(15,[8],[1.0])|      (9,[5],[1.0])|      (8,[5],[1.0])|(2,
[0],[1.0])|      (2,[0],[1.0])|      (3,[],[])|(29,[0,2,5,14,21,...| 582.7858526373045|
|      DRC01|      5.92|      Regular|      0.019308607|Soft Drinks| 49.0692|
      OUT017|      2007|      Medium|      Tier 2|Supermarket Type
1|      1478.076|      673.0|      1.0|      8.0|      2.0|
      3.0|      0.0|      1.0|      0.0| (1553,[673],[1.
0])|      (4,[1],[1.0])|(15,[8],[1.0])|      (9,[2],[1.0])|      (8,[3],[1.0])|(2,
[0],[1.0])|      (2,[1],[1.0])|(3,[0],[1.0])|(29,[0,2,5,14,21,...| 929.3680572902323|
|      DRC12|      17.85|      Low Fat|      0.03781972|Soft Drinks|191.6188|
      OUT035|      2004|      Small|      Tier 2|Supermarket Type
1|      2475.4444|      1498.0|      0.0|      8.0|      1.0|
      2.0|      1.0|      1.0|      0.0| (1553,[1498],[1.
0])|      (4,[0],[1.0])|(15,[8],[1.0])|      (9,[1],[1.0])|      (8,[2],[1.0])|(2,
[1],[1.0])|      (2,[1],[1.0])|(3,[0],[1.0])|(29,[0,1,5,14,21,...| 3030.7345765579985|
|      DRC12|      17.85|      Low Fat|      0.037826873|Soft Drinks|189.7188|
      OUT046|      1997|      Small|      Tier 1|Supermarket Type
1|      2285.0256|      1498.0|      0.0|      8.0|      6.0|
      6.0|      1.0|      2.0|      0.0| (1553,[1498],[1.
0])|      (4,[0],[1.0])|(15,[8],[1.0])|      (9,[6],[1.0])|      (8,[6],[1.0])|(2,
[1],[1.0])|      (2,[],[])|(3,[0],[1.0])|(29,[0,1,5,14,21,...| 3023.129033573159|
|      DRC12|      17.85|      Low Fat|      0.038040837|Soft Drinks|189.1188|
      OUT017|      2007|      Medium|      Tier 2|Supermarket Type
1|      3237.1196|      1498.0|      0.0|      8.0|      2.0|
      3.0|      0.0|      1.0|      0.0| (1553,[1498],[1.
0])|      (4,[0],[1.0])|(15,[8],[1.0])|      (9,[2],[1.0])|      (8,[3],[1.0])|(2,
[0],[1.0])|      (2,[1],[1.0])|(3,[0],[1.0])|(29,[0,1,5,14,21,...| 2991.8000936798494|
|      DRC13|      8.26|      Regular|      0.032573725|Soft Drinks| 125.073|
      OUT018|      2009|      Medium|      Tier 3|Supermarket Type
2|      985.384|      1337.0|      1.0|      8.0|      5.0|
      5.0|      0.0|      0.0|      3.0| (1553,[1337],[1.
0])|      (4,[1],[1.0])|(15,[8],[1.0])|      (9,[5],[1.0])|      (8,[5],[1.0])|(2,
[0],[1.0])|      (2,[0],[1.0])|      (3,[],[])|(29,[0,2,5,14,21,...| 1764.2353427575733|
+-----+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 20 rows

```

Evaluate Model

```
In [ ]: from pyspark.ml.evaluation import RegressionEvaluator
```

```
In [ ]: errors = ["r2", "rmse", "mse", "mae"]
name = ["R-Square or Accuracy", "Root Mean Square Error", "Mean Square Error", "Mean Abs

for i in range(len(errors)):
    eval = RegressionEvaluator(predictionCol="prediction", labelCol='Item_Outlet_Sales', m
    print("The {} of Model is {}".format(name[i],eval.evaluate(result)))
```

```

The R-Square or Accuracy of Model is 0.5609324399455548
The Root Mean Square Error of Model is 1146.154277794764
The Mean Square Error of Model is 1313669.6285072374
The Mean Absolute Error of Model is 854.720185337692

```


